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Difference equations in a general setting

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Abstract

For a square matrix A , the difference equation $y_{k+1} = y_k A$ generates a sequence of vectors y_0, y_1, \dots . The long run behavior of this sequence is described in terms of a group obtained from A, A^2, \dots . The results are given in a topological semigroup setting so they can also be applied to more general difference equations. © 2000 Elsevier Science Inc. All rights reserved.

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Let A be an $n \times n$ matrix. The classical difference equation

$$y_{k+1} = y_k A,$$

where y_0 is a vector, arises in numerous applied problems. (See [6].)

The solution to this difference equation need not converge. In those cases, and for other reasons, it is sometimes useful to consider a difference equation

$$S_{k+1} = S_k A,$$

where S_0 is a set of vectors and $S_{k+1} = \{x A : x \in S_k\}$. For example, if

$$S_0 = \text{convex}\{(1, 0, 0), (0, 1, 0), (0, 0, 1)\},$$

the set of all probability distributions, it can be shown that for

$$A = \begin{bmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ 1 & 0 & 0 \\ 1 & 0 & 0 \end{bmatrix}$$

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S_0, S_1, \dots converges, in the Hausdorff metric, to $\text{convex}\{(1, 0, 0), (0, \frac{1}{2}, \frac{1}{2})\}$. This occurs even though the classical difference equation $y_{k+1} = y_k A$ need not converge for particular probability distributions. (See [2,3] for more details.)

An even more general way to study difference equations is to consider

$$S_{k+1} = S_k M,$$

where S_0 is a set of vectors, M is a set of $n \times n$ matrices, and $S_{k+1} = \{xA : x \in S_k \text{ and } A \in M\}$. Barnsley [1] uses this approach to construct fractals and Hartfiel [4] uses it to study Markov chains with imprecise data.

What we intend to show in this paper is that the behavior of solutions to difference equations, in these various settings, can be described using a group obtained from the sequence A, A^2, \dots or M, M^2, \dots .

To do this, let p be an $n \times n$ real matrix or a set of such matrices. Let W be a subset of \mathbb{R}^n . We consider

$$W, Wp, Wp^2, \dots$$

By using topological semigroups, we intend to describe the long run behavior of the sequence.

Results. Let S be a compact topological semigroup (S is a Hausdorff space and multiplication in S is continuous) whose topology is determined by a continuous metric d . Some examples of such follow.

- (i) The semigroups of the $n \times n$ stochastic matrices and the $n \times n$ substochastic matrices: Here we use the 1-norm for vectors and the induced matrix norm $\|\cdot\|_1$, under left multiplication by those vectors,

$$\|A\|_1 = \max_{x \neq 0} \frac{\|xA\|_1}{\|x\|_1}.$$

Recall that

$$\|A\|_1 = \max_i \sum_{k=1}^n |a_{ik}|$$

and

$$d(A, B) = \|A - B\|_1.$$

- (ii) The semigroup of $n \times n$ orthogonal matrices: Using the 2-norm for vectors and the induced matrix norm $\|\cdot\|_2$ we have

$$d(A, B) = \|A - B\|_2.$$

- (iii) Special 0-pattern matrices: Let $\mathbb{R}^{n \times n}$ be the set of $n \times n$ ($n > 1$) real matrices with the induced norm $\|\cdot\|_\infty$ for left multiplication by vectors. Thus,

$$\|A\|_\infty = \max_j \sum_{k=1}^n |a_{kj}|$$

and

$$d(A, B) = \|A - B\|_\infty.$$

Let S be the subset of $\mathbb{R}^{n \times n}$ defined by

$$S = \left\{ A: A = \begin{bmatrix} B & 0 \\ b & 1 \end{bmatrix} \text{ and } \|A\|_\infty \leq 1 \right\}.$$

Then S is a compact topological semigroup.

- (iv) Finite generating sets: There are numerous variants of the semigroup given in (iii). We look at one of these.

Barnsley uses the affine maps

$$x_{k+1} = x_k \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix} + (1, 1); \quad x_{k+1} = x_k \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix} + (1, 50);$$

$$x_{k+1} = \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix} x_k + (50, 50)$$

to construct the Sierpinski triangle. (Other fractals require other affine maps.)

For a given triangle T_0 , Barnsley computes

$$\begin{aligned} T_{k+1} &= \left(T_k \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix} + (1, 1) \right) \cup \left(T_k \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix} + (1, 50) \right) \\ &\quad \cup \left(T_k \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix} + (50, 50) \right) \\ &= \left(T_k \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix} + (1, 1) \right) = \left\{ x \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix} + (1, 1): x \in T_k \right\}, \text{ etc.} \end{aligned}$$

and shows $T_0 \cup T_1 \cup \dots$ gives the Sierpinski triangle.

To put this construction into our difference equation format set

$$\begin{aligned} A_1 &= \begin{bmatrix} 0.5 & 0 & 0 \\ 0 & 0.5 & 0 \\ 1 & 1 & 1 \end{bmatrix}, \quad A_2 = \begin{bmatrix} 0.5 & 0 & 0 \\ 0 & 0.5 & 0 \\ 1 & 50 & 1 \end{bmatrix}, \\ A_3 &= \begin{bmatrix} 0.5 & 0 & 0 \\ 0 & 0.5 & 0 \\ 50 & 50 & 1 \end{bmatrix} \end{aligned}$$

and $y_k = (x_k, 1)$. Thus, using these homogeneous coordinates, we have

$$S_{k+1} = S_k A_1 \cup S_k A_2 \cup S_k A_3 = S_k \{A_1, A_2, A_3\},$$

where $S_k = \{(x, 1): x \in T_k\}$. In this notation

$$S_0 \cup S_1 \cup \dots$$

gives the Sierpinski triangle.

To obtain a semigroup, let $N = \{I, A_1, A_2, A_3\}$. Define

$$\begin{aligned} N^2 &= \{C_{i_1}C_{i_2} : C_{i_1}, C_{i_2} \in N\}, \\ N^3 &= \{C_{i_1}C_{i_2}C_{i_3} : C_{i_1}, C_{i_2}, C_{i_3} \in N\}, \\ &\dots \end{aligned}$$

Then

$$S = N \cup N^2 \cup N^3 \cup \dots$$

is a semigroup.

For the norm we have

$$\|A\| = \|DAD^{-1}\|_\infty,$$

where

$$D = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0.001 \end{bmatrix},$$

to assure $\|C\| \leq 1$ for all $C \in N$ and thus for all $C \in S$. Using this norm on $\mathbb{R}^{3 \times 3}$, \overline{S} , the closure of S , is a compact semigroup.

Finally, if

$$\begin{aligned} C_k &= S_0 \cup \dots \cup S_k \quad \text{then} \\ C_{k+1} &= C_k N \end{aligned}$$

and C_k approaches the Sierpinski triangle. This puts fractal construction of Barnsley in our setting.

- (v) A semigroup of compact sets: Given a norm $\|\cdot\|$ on the set $\mathbb{R}^{n \times n}$ of $n \times n$ matrices and that $d(A, B) = \|A - B\|$, let S be any compact semigroup of $\mathbb{R}^{n \times n}$. Define

$$H(S) = \{X : X \text{ is a nonempty compact set in } S\}.$$

Using the product of sets, $H(S)$ is a semigroup. Further, with the Hausdorff metric, $H(S)$ is a compact semigroup (see Barnsley [1]).

The semigroup $H(S)$, where S is the set of stochastic matrices, is used in Hartfiel [4], a study concerning imprecise transition matrices, for Markov chains.

We now develop the general results of the paper. Notationally, we let S be a compact topological semigroup with continuous metric d and $p \in S$. Consider the sequence p, p^2, p^3, \dots . It is shown in [5] that the set G of limit points of this sequence is a group.

Before giving the major work of this paper, we prove two results about G .

Lemma 1. Define $\pi : G \rightarrow G$ by $\pi(g) = gp$. Then π is an isomorphism.

Proof. To show that $\pi : G \rightarrow G$, let $g \in G$. Then there is a sequence p^{i_1}, p^{i_2}, \dots which converges to g . Since multiplication is continuous in a topological semig-

roup, $p^{i_1+1}, p^{i_2+1}, \dots$ converges to gp . Thus, gp is a limit point of p, p^2, \dots and so $gp \in G$.

To show π is onto, let $g \in G$. Then, g is the limit of some sequence p^{i_1}, p^{i_2}, \dots . Consider $p^{j_1-1}, p^{j_2-1}, \dots$ and take a convergent subsequence of it say $p^{j_1-1}, p^{j_2-1}, \dots$. Suppose this sequence converges to $\bar{g} \in G$. Then $\bar{g}p$ is the limit point of p^{j_1}, p^{j_2}, \dots which is g . Thus, $\bar{g}p = g$ and $\pi: G \rightarrow G$ is onto.

Finally, to show that π is one-to-one, suppose that $g, \bar{g} \in G$ and that $gp = \bar{g}p$. Let e be the identity element of G . Then $g(ep) = \bar{g}(ep)$. Since $\pi: G \rightarrow G$, $ep \in G$ and thus has an inverse there. Hence, $g = \bar{g}$ which shows that p is one-to-one. \square

Define the coefficient of contraction (or contractive factor) for p as

$$\mathcal{T}_p = \sup_{x, y} \frac{d(xp, yp)}{d(x, y)},$$

where the sup is over all distinct $x, y \in \mathbb{R}^n$. We are interested in those semigroups in which $\mathcal{T}_p \leq 1$. For the stochastic matrix semigroup, the substochastic matrix semigroup, the orthogonal matrix semigroup, and the semigroups of (iii) and (iv), $\mathcal{T}_p \leq \|p\|$ and $\|p\| \leq 1$. Further, for $H(S)$ defined in (v), $\mathcal{T}_p \leq 1$ is shown in [4].

Using this notation we give a condition that assures G is finite. Later we will show that G determines the behavior of difference equations, as described in the introduction, and when G is finite that behavior is more easily calculated.

Lemma 2. Suppose $\mathcal{T}_p \leq 1$. Then G is finite if and only if there is a positive integer s and an element $g \in G$ such that $gp^s = g$.

Proof. By Lemma 1, the direct implication is clear. Thus, we need only argue the converse implication. For this, we let g denote an element as described in the converse.

Set $g_0 = g$, $g_1 = gp$, \dots , $g_{s-1} = gp^{s-1}$, $gp^s = g$, where s is the smallest such integer. We show $G = \{g_0, \dots, g_{s-1}\}$. The proof is by contradiction.

Suppose there is a $g_s \in G$ and $g_s \neq g_k$ for $k = 0, \dots, s-1$. Let $\epsilon = \frac{1}{3} \min d(g_i, g_j)$. (The minimum is taken over i, j where $0 \leq i \neq j \leq s$.) Let p^{i_1}, p^{i_2}, \dots be a subsequence of p^1, p^2, \dots which converges to g . Let $K > 0$ such that if $k > K$ then $d(p^{i_k}, g) < \epsilon$. Then if t is a positive integer and $t = sq + r$ where $0 \leq r < s$, $d(p^{i_k+t}, g_r) = d(p^{i_k+t}, gp^r) = d(p^{i_k} p^{sq+r}, gp^{sq+r}) \leq d(p^{i_k}, g) < \epsilon$. Thus, for $i > i_K$, $d(p^i, g_s) > \epsilon$ and so g_s is not a limit point of p^1, p^2, \dots . Hence $g_s \notin G$, a contradiction from which the result follows. \square

For the major work we let W be a compact subset of \mathbb{R}^n and p an $n \times n$ matrix (in a compact topological semigroup S in $\mathbb{R}^{n \times n}$) or a compact set of $n \times n$ matrices (in a compact topological semigroup as described in (v)).

Consider the sequence of compact sets Wp, Wp^2, \dots . Define the limit set W_∞ as

$$W_\infty = \{s \in W : s \text{ is a limit point of } w_1p, w_2p^2, \dots \text{ where } w_1, w_2, \dots \text{ are in } W\}. \quad (1)$$

It is easily seen that W_∞ is compact. We now show that it is G that determines W_∞ from W . This description is below.

Theorem 1. $W_\infty = \bigcup_{g \in G} Wg$.

Proof. Let $u \in W_\infty$. Then there is a sequence $w_{i_1}p^{i_1}, w_{i_2}p^{i_2}, \dots$ which converges to u . Let w_{j_1}, w_{j_2}, \dots be a subsequence of w_{i_1}, w_{i_2}, \dots that converges, say to w . Since W is compact, $w \in W$. Let p^{r_1}, p^{r_2}, \dots be a subsequence of p^{i_1}, p^{i_2}, \dots which converges to say g . Thus, $g \in G$. Now, $u = \lim_{k \rightarrow \infty} w_{r_k}p^{r_k} = wg \in Wg$. Hence $W_\infty \subseteq \bigcup_{g \in G} Wg$.

Conversely, let $u \in \bigcup_{g \in G} Wg$. Then, $u \in Wg$ for some $g \in G$. Thus, $u = wg$ for some $w \in W$. Since $g \in G$, there is a sequence p^{i_1}, p^{i_2}, \dots which converges to g . Thus, $wp^{i_1}, wp^{i_2}, \dots$ converges to wg . And hence, since $wg = u$, $u \in W_\infty$. So, $\bigcup_{g \in G} Wg \subseteq W_\infty$. \square

Two examples of this theorem follow.

Example 1. Chi [2] computed the limit set for

$$W = \text{convex}\{(1, 0, 0, 0), (0, 1, 0, 0), (0, 0, 1, 0), (0, 0, 0, 1)\}$$

and

$$p = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & \frac{1}{2} & \frac{1}{2} \\ 0 & 0 & 0 & 1 \end{bmatrix},$$

using the 1-norm.

To use our theorem for this computation, we factor

$$p = PDP^{-1}$$

where $D = \text{diag}(1, 1, -1, \frac{1}{2})$ from which we get the limit points

$$g_1 = P \text{diag}(1, 1, -1, 0)P^{-1} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix},$$

$$g_2 = P \text{diag}(1, 1, 1, 0)P^{-1} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix}.$$

From this we have

$$\begin{aligned} W_\infty &= Wg_1 \cup Wg_2 \\ &= \text{convex}\{(1, 0, 0, 0), (0, 1, 0, 0), (0, 0, 0, 1)\}. \end{aligned}$$

Example 2. Let $X_0 = \text{convex}\{(0, 0), (1, 0), (0, 1)\}$ and consider

$$X_{k+1} = X_k A + (1, 1),$$

where

$$A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}.$$

We study the behavior of X_0, X_1, \dots

Using homogeneous coordinates we can write this as

$$W_{k+1} = W_k p,$$

where

$$p = \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 1 & 1 & 1 \end{bmatrix}$$

and

$$W_k = \{(x, 1) : x \in X_k\}.$$

Note that there is a nonsingular matrix R such that $P = R \text{diag}(1, 1, -1)R^{-1}$ so we use the norm $\|x\| = \|xR\|_\infty$ to get $\|p\| \leq 1$.

Now

$$\begin{aligned} p &= \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 1 & 1 & 1 \end{bmatrix}, & p^2 &= \begin{bmatrix} -1 & 0 & 0 \\ 0 & -1 & 0 \\ 2 & 0 & 1 \end{bmatrix}, \\ p^3 &= \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 1 & -1 & 1 \end{bmatrix}, & p^4 &= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}. \end{aligned}$$

So $G = \{p, p^2, p^3, p^4\}$. Thus,

$$\begin{aligned} W_\infty &= Wp \cup Wp^2 \cup Wp^3 \cup Wp^4 \\ &= \text{convex}\{(1, 0, 1), (1, 1, 1), (2, 1, 1)\} \\ &\quad \cup \text{convex}\{(1, 0, 1), (2, 0, 1), (2, -1, 1)\} \\ &\quad \cup \text{convex}\{(1, 0, 1), (1, -1, 1), (0, -1, 1)\} \\ &\quad \cup \text{convex}\{(0, 0, 1), (1, 0, 1), (0, 1, 1)\} \end{aligned}$$

as shown in Fig. 1.

Note that π permutes the triangles in Fig. 1. And, dropping the third coordinate in the vectors in W_∞ (or projecting the graph into the xy -plane) we see the behavior of X_0, X_1, \dots

We now show some further details of how the sequence W, Wp, Wp^2, \dots tends to W_∞ .

Theorem 2. $Wp^k \rightarrow W_\infty$ in the sense that if \mathcal{O} is an open set containing W_∞ then there is a K such that if $k \geq K$ then $Wp^k \subseteq \mathcal{O}$.

Proof. The proof is by contradiction. Suppose there is an open set \mathcal{O} about W_∞ , such that $Wp^j \not\subseteq \mathcal{O}$ for $j = i_1, i_2, \dots$. Choose $w_{i_1}p^{i_1} \in Wp^{i_1}, w_{i_2}p^{i_2} \in Wp^{i_2}, \dots$ so that $w_{i_1}p^{i_1} \notin \mathcal{O}, w_{i_2}p^{i_2} \notin \mathcal{O}, \dots$. Let w_{j_1}, w_{j_2}, \dots be a subsequence of w_{i_1}, w_{i_2}, \dots which converges to say w and p^{r_1}, p^{r_2}, \dots a subsequence of p^{j_1}, p^{j_2}, \dots that converges to say $\bar{g} \in G$. Then $w_{r_1}p^{r_1}, w_{r_2}p^{r_2}, \dots$ converges to $w\bar{g}$. But $w\bar{g} \in W_\infty$ which is a contradiction since $w\bar{g} \notin \mathcal{O}$. Thus, the result follows. \square

The following two corollaries use the Hausdorff metric Δ and the following dilation result given in [1]: Let $\epsilon > 0$. If $A \subseteq B + \epsilon$ and $B \subseteq A + \epsilon$ then $\Delta(A, B) \leq \epsilon$. (Here $B + \epsilon = \{s \in S : d(s, b) \leq \epsilon \text{ for some } b \in B\}$.)

Corollary 1. If $G = \{e\}$ then

$$\Delta(Wp^k, W_\infty) \rightarrow 0 \quad \text{as } k \rightarrow \infty.$$

Proof. The proof follows by showing that for any $\epsilon > 0$ there is a K such that

$$d(wp^k, we) < \epsilon \quad \text{for all } w \in W \text{ and } k \geq K.$$

The proof of this result is argued by contradiction as in the proof of the preceding theorem. \square

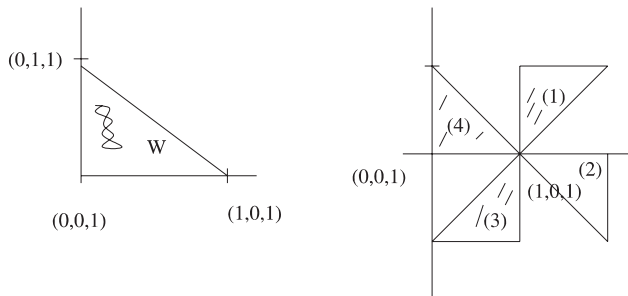


Fig. 1. Graph of W and Wp, Wp^2, Wp^3, Wp^4 labeled (1), (2), (3), (4), respectively.

A result that appears in [2,3] follows.

Corollary 2. *If $Wp \subseteq W$ then $\Delta(Wp^k, W_\infty) \rightarrow 0$ as $k \rightarrow \infty$.*

Proof. It is clear that $Wp^{k+1} \subseteq Wp^k$ for all k . It follows that $W_\infty \subseteq Wp^k$ for all k . And, by Theorem 3 we know that given any $\epsilon > 0$ there is a k such that $Wp^k \subseteq W_\infty + \epsilon$. Thus, the corollary follows. \square

A theorem giving a more exacting description of the behavior of the sequence Wp, Wp^2, \dots utilizes the following lemma.

Lemma 3. *Given $\epsilon > 0$ and a $g \in G$ there is a $k > 0$ such that*

$$d(wp^k, wg) < \epsilon \quad \text{for all } w \in W.$$

Proof. The proof is by contradiction. Thus, suppose there is an $\epsilon > 0$ and a $g \in G$ so that for each k there is a $\bar{w} \in W$ so that

$$d(\bar{w}p^k, \bar{w}g) \geq \epsilon.$$

Now let p^{i_1}, p^{i_2}, \dots be a convergent sequence which converges to g . Now for each i_k , let $w_{i_k} \in W$ be such that

$$d(w_{i_k}p^{i_k}, w_{i_k}g) \geq \epsilon.$$

Let w_{j_1}, w_{j_2}, \dots be a subsequence of w_{i_1}, w_{i_2}, \dots which converges to say \bar{w} . Since d is continuous, it follows that

$$d(\bar{w}g, \bar{w}g) \geq \epsilon$$

a contradiction from which the result follows. \square

Theorem 3. *Suppose $\mathcal{T}_p \leq 1$. Given $\epsilon > 0$ and $g \in G$ there is a K so that if $k \geq 0$ then*

$$\Delta(Wp^{K+k}, Wgp^k) < \epsilon.$$

Proof. By Lemma 3 we know that there is a K such that

$$d(wp^K, wg) < \epsilon \quad \text{for all } w \in W.$$

Thus, since $\mathcal{T}_p \leq 1$,

$$d(wp^{K+k}, wgp^k) < \epsilon \quad \text{for all } w \in W.$$

This is sufficient to show that

$$\Delta(Wp^{K+k}, Wgp^k) < \epsilon. \quad \square$$

This theorem says that for large k , Wp^k is close, in the Hausdorff metric, to some Wg , the g perhaps changing with k . (Note, by Lemma 1, $gp^k \in G$.)

The following example can be helpful.

Example 3. Let

$$p = \left\{ \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \frac{1}{2} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right\}.$$

Using the 1-norm, as in (v), $\|p\| \leq 1$.

Consider the difference equation

$$S_{k+1} = S_k p,$$

where $W = \{(1, 0)\}$. A few calculations,

$$\begin{aligned} p &= \left\{ \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \frac{1}{2} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right\}, \\ p^2 &= \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \frac{1}{2} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \frac{1}{4} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right\}, \\ p^3 &= \left\{ \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \frac{1}{2} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \frac{1}{4} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \frac{1}{8} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right\}, \\ p^4 &= \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \frac{1}{2} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \frac{1}{4} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \frac{1}{8} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \frac{1}{16} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right\}, \\ &\dots \end{aligned}$$

shows that $G = \{p_0, p_1\}$ where

$$\begin{aligned} p_0 &= \left\{ \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \frac{1}{2} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \frac{1}{4} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \frac{1}{8} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \dots, 0 \right\}, \\ p_1 &= \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \frac{1}{2} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \frac{1}{4} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \frac{1}{8} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \dots, 0 \right\}. \end{aligned}$$

(0 is added to assure p_0 and p_1 are compact.)

Thus,

$$\begin{aligned} W_\infty &= Wp_0 \cup Wp_1 \\ &= \left\{ (0, 1), \left(\frac{1}{2}, 0 \right), \left(0, \frac{1}{4} \right), \left(\frac{1}{8}, 0 \right), \dots \right\} \\ &\quad \cup \left\{ (1, 0), \left(0, \frac{1}{2} \right), \left(\frac{1}{4}, 0 \right), \left(0, \frac{1}{8} \right), \dots \right\}. \end{aligned}$$

The first few iterates of this set are shown in Fig. 2 where squares are used for Wp_0 and circles for Wp_1 . In viewing W, Wp, \dots we see that odd iterates tend to Wp_0 while even iterates tend to Wp_1 . Note also that if we replace $\frac{1}{2}$ by w , $0 \leq w < 1$, in the definition of p , then W_∞ is infinite except at $w = 0$.

An example requiring no computation follows.

Example 4. Let P be an $n \times n$ positive (entrywise) matrix and

$$p = \{A: A \text{ is an } n \times n \text{ doubly stochastic matrix and } A \geq P(\text{entrywise})\}.$$

Using the techniques in [4] it can be shown that p, p^2, \dots converges to $\{J\}$ where J is an $n \times n$ matrix with $\frac{1}{n}$ as each entry. Thus, if W is any compact set of stochastic vectors then

$$Wp, Wp^2, \dots$$

converges to $W\{J\}$.

The following corollary, in special situations, appears in [2–4] as well as in other writings involving difference equations.

Corollary 3. $W_\infty p = \bigcup_{g \in G} W(gp) = W_\infty$.

Proof. Follows from Theorem 1 and Lemma 1. \square

Corollary 3 indicates that although π may contract W , it does not contract W_∞ . So, for example, if W_∞ is a fractal, π will not contract it. More specifically we have the following result about W_∞ which we feel is new.

Theorem 4. If $\mathcal{T}_p \leq 1$ then π is an isometry on W_∞ .

Proof. Let $\bar{x}, \bar{y} \in S$. Set $x = \bar{x}g, y = \bar{y}g$ where $g \in G$. Let p^{i_1}, p^{i_2}, \dots be a sequence that converges to g .

Note that

$$0 \leq d(\bar{x}p^{i_k}, \bar{y}p^{i_k}) - d(\bar{x}p^{i_k}p, \bar{y}p^{i_k}p) \leq d(\bar{x}p^{i_k}, \bar{y}p^{i_k}) - d(\bar{x}p^{i_{k+1}}, \bar{y}p^{i_{k+1}}).$$

Taking the limit as $k \rightarrow \infty$ yields

$$0 \leq d(\bar{x}g, \bar{y}g) - d(\bar{x}gp, \bar{y}gp) \leq d(\bar{x}g, \bar{y}g) - d(\bar{x}g, \bar{y}g),$$

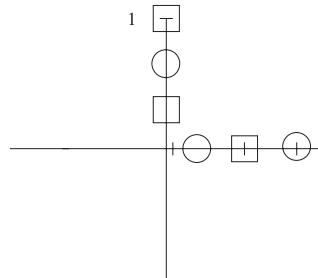


Fig. 2. The graph of W_∞ .

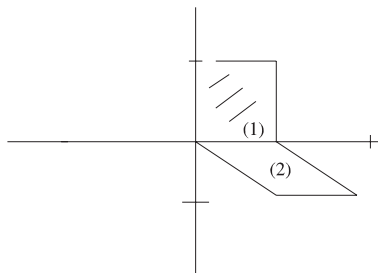


Fig. 3. Graph of W_∞ . Here Wp , Wp^2 are labeled (1) and (2), respectively.

so

$$d(\bar{x}gp, \bar{y}gp) = d(\bar{x}g, \bar{y}g).$$

Now suppose that $x = \bar{x}g_1$, $y = \bar{y}g_2 \in S_\infty$ where $g_1, g_2 \in G$. Then, using the special case above,

$$d(\bar{x}g_1p, (\bar{y}g_2g_1^{-1})g_1p) = d(\bar{x}g_1, (\bar{y}g_2g_1^{-1})g_1).$$

So

$$d(\bar{x}g_1p, \bar{y}g_2p) = d(\bar{x}g_1, \bar{y}g_2).$$

The result follows. \square

It is interesting to note that the isometry depends on the metric d used for \mathcal{T}_p .

Example 5. Let

$$p = \begin{bmatrix} 1 & 0 \\ 1 & -1 \end{bmatrix}.$$

Since the eigenvalues of p are 1 and -1 , as in Example 2, there is a norm $\|\cdot\|$ so that $\|p\| \leq 1$.

Let $W = \text{convex}\{(0, 0), (1, 0), (0, 1), (1, 1)\}$, a square. Since $p^2 = I$ it follows that

$$W_\infty = Wp \cup Wp^2$$

as shown in Fig. 3.

Note that $(Wp)p = Wp^2$ while $(Wp^2)p = Wp$ thus multiplying by p interchanges the parallelograms labeled (1) and (2). Thus, it is clear that π does not preserve distances between points in W , using the 2-norm.

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